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LAMPIRAN

Lampiran 1. Data Status Gizi Balita di Kabupaten Gowa Tahun 2022

No.	Jenis Kelamin	Usia (Bulan)	Berat Badan (Kg)	Panjang Badan (Cm)	Status Gizi
1	Laki-laki	34	13.2	93.0	Gizi Baik
2	Perempuan	25	12.8	84.0	Gizi Lebih
3	Laki-laki	36	13.5	90.5	Gizi Baik
4	Laki-laki	53	15.0	100.0	Gizi Baik
5	Perempuan	45	13.2	99.0	Gizi Baik
6	Perempuan	51	14.4	99.0	Gizi Baik
7	Perempuan	45	13.5	96.0	Gizi Baik
8	Perempuan	22	7.7	75.5	Gizi Kurang
9	Perempuan	22	9.4	79.7	Gizi Baik
10	Laki-laki	5	6.0	61.0	Gizi Baik
11	Perempuan	22	10.2	80.0	Gizi Baik
12	Laki-laki	49	15.0	100.0	Gizi Baik
13	Perempuan	56	14.0	98.0	Gizi Baik
14	Perempuan	53	13.9	101.0	Gizi Baik
15	Perempuan	58	18.8	102.1	Gizi Lebih
16	Perempuan	9	5.8	64.5	Gizi Kurang
17	Laki-laki	42	10.5	90.0	Gizi Kurang
18	Perempuan	26	9.0	84.0	Gizi Kurang
19	Perempuan	26	9.5	87.0	Gizi Kurang
20	Laki-laki	50	18.0	101.0	Gizi Lebih
21	Laki-laki	36	12.0	95.0	Gizi Kurang
22	Laki-laki	8	7.0	70.0	Gizi Kurang
23	Laki-laki	36	10.7	90.2	Gizi Kurang
24	Perempuan	50	16.4	95.3	Gizi Lebih
25	Laki-laki	14	10.0	72.0	Gizi Lebih
26	Perempuan	52	17.8	102.0	Gizi Lebih
27	Perempuan	11	8.7	68.0	Gizi Lebih
28	Laki-laki	17	10.7	75.0	Gizi Lebih
29	Laki-laki	42	16.6	98.0	Gizi Lebih
30	Laki-laki	26	8.0	82.0	Gizi Kurang
31	Laki-laki	55	15.3	93.0	Gizi Lebih

32	Laki-laki	58	18.9	105.0	Gizi Lebih
33	Laki-laki	47	11.7	95.0	Gizi Kurang
34	Laki-laki	6	7.7	60.6	Gizi Lebih
35	Laki-laki	49	12.0	99.0	Gizi Kurang
36	Laki-laki	12	9.8	70.3	Gizi Lebih
37	Laki-laki	60	15.5	110.0	Gizi Kurang
38	Laki-laki	58	13.9	113.0	Gizi Kurang
39	Perempuan	12	7.0	72.0	Gizi Kurang
40	Laki-laki	34	13.5	86.8	Gizi Lebih
41	Laki-laki	54	11.0	94.0	Gizi Kurang
42	Laki-laki	34	14.0	90.0	Gizi Lebih
43	Laki-laki	45	11.8	95.0	Gizi Kurang
44	Laki-laki	49	14.7	108.0	Gizi Kurang
45	Perempuan	30	13.0	85.0	Gizi Lebih
46	Laki-laki	45	11.6	93.0	Gizi Kurang
47	Perempuan	3	2.3	47.0	Gizi Kurang
48	Perempuan	22	7.7	78.0	Gizi Kurang
49	Laki-laki	54	12.7	99.0	Gizi Kurang
50	Laki-laki	28	9.5	83.5	Gizi Kurang
⋮	⋮	⋮	⋮	⋮	⋮
17600	Laki-laki	14	10.2	73.2	Gizi Lebih

Sumber: Dinas Kesehatan, 2022

Lampiran 2. Titik Knot dan Nilai GCV untuk 1 Titik Knot

Nilai GCV untuk Variabel Usia (x_1) disetiap Orde

No	Orde Linier		Orde Kuadratik		Orde Kubik	
	k_{11}	Nilai GCV	k_{11}	Nilai GCV	k_{11}	Nilai GCV
1	9	0.39794	3	0.39985	2	0.40026
2	10	0.39797	2	0.39985	3	0.40028
3	11	0.39798	4	0.39985	4	0.40029
4	12	0.39799	5	0.39985	5	0.40031
5	13	0.39801	6	0.39986	6	0.40032
6	14	0.39803	7	0.39987	7	0.40034
7	15	0.39807	8	0.39988	8	0.40036
8	16	0.39813	9	0.39990	9	0.40038
9	8	0.39819	10	0.39993	10	0.40040
10	17	0.39825	11	0.39995	11	0.40042
11	18	0.39847	12	0.39998	12	0.40044
12	7	0.39848	13	0.40002	13	0.40046
13	19	0.39866	14	0.40005	14	0.40048
14	20	0.39888	15	0.40009	15	0.40050
15	6	0.39896	16	0.40013	16	0.40052
16	21	0.39909	17	0.40018	17	0.40054
17	22	0.39932	18	0.40022	18	0.40056
18	23	0.39951	19	0.40026	19	0.40058
19	24	0.39969	20	0.40030	20	0.40060
20	5	0.39972	21	0.40034	21	0.40062
21	25	0.39983	22	0.40038	22	0.40063
22	26	0.39994	23	0.40042	23	0.40065
23	27	0.40003	24	0.40045	24	0.40067
⋮	⋮	⋮	⋮	⋮	⋮	⋮
58	59	0.40117	59	0.40117	59	0.40117

Sumber: Data olah, 2023

Nilai GCV untuk Variabel Berat Badan (x_2) disetiap Orde

No	Orde Linier		Orde Kuadratik		Orde Kubik	
	k_{21}	Nilai GCV	k_{21}	Nilai GCV	k_{21}	Nilai GCV
1	15	0.37795	12	0.38628	5	0.39060
2	15.1	0.37798	12.1	0.38628	4.9	0.39060
3	14.95	0.37800	11.9	0.38629	5.1	0.39060
4	14.9	0.37808	12.2	0.38630	4.8	0.39060
5	15.2	0.37808	11.8	0.38633	4.7	0.39060
6	15.3	0.37825	12.3	0.38634	5.2	0.39060
7	14.8	0.37830	11.75	0.38635	4.6	0.39061
8	15.4	0.37846	12.35	0.38637	5.3	0.39061
9	14.7	0.37861	11.7	0.38637	4.5	0.39061
10	15.5	0.37874	12.4	0.38640	5.4	0.39061
11	14.6	0.37900	11.6	0.38643	4.4	0.39061
12	15.6	0.37908	12.5	0.38649	5.5	0.39062
13	14.51	0.37942	11.5	0.38650	4.3	0.39062
14	14.5	0.37947	12.53	0.38652	5.57	0.39062
15	15.7	0.37950	11.46	0.38654	5.6	0.39062
16	15.8	0.37995	12.55	0.38654	4.2	0.39062
17	14.4	0.37997	11.45	0.38654	4.1	0.39063
18	15.9	0.38052	11.4	0.38659	5.7	0.39063
19	14.3	0.38052	12.6	0.38660	4	0.39064
20	14.25	0.38082	12.65	0.38667	5.8	0.39064
21	14.2	0.38114	11.3	0.38668	3.9	0.39064
22	16	0.38118	12.7	0.38674	3.8	0.39065
23	14.15	0.38146	11.2	0.38678	5.9	0.39066
⋮	⋮	⋮	⋮	⋮	⋮	⋮
250	3.7	0.41736	30	0.41716	19.9	0.41549

Sumber: Data olah, 2023

Nilai GCV untuk Variabel Panjang Badan (x_3) disetiap Orde

No	Orde Linier		Orde Kuadratik		Orde Kubik	
	k_{31}	Nilai GCV	k_{31}	Nilai GCV	k_{31}	Nilai GCV
1	74.5	0.40006	60.5	0.40005	47	0.40013
2	74.4	0.40006	60.3	0.40005	48	0.40013
3	74.7	0.40007	60.2	0.40005	49	0.40013
4	74.6	0.40007	60.6	0.40005	49.3	0.40014
5	74.8	0.40007	60.1	0.40005	49.5	0.40014
6	74.3	0.40007	60.7	0.40005	49.7	0.40014
7	74.2	0.40007	60	0.40005	50	0.40014
8	74.9	0.40007	60.8	0.40005	50.3	0.40014
9	75.1	0.40007	59.9	0.40005	50.5	0.40014
10	74.1	0.40007	60.9	0.40005	50.8	0.40014
11	75.2	0.40007	59.8	0.40005	51	0.40014
12	75	0.40007	61	0.40005	51.1	0.40014
13	74	0.40007	59.7	0.40005	51.2	0.40014
14	75.3	0.40007	61.1	0.40005	51.3	0.40014
15	75.4	0.40007	59.6	0.40005	51.4	0.40014
16	73.9	0.40007	61.2	0.40005	51.5	0.40014
17	75.5	0.40007	59.5	0.40005	51.7	0.40014
18	73.8	0.40007	61.3	0.40005	52	0.40014
19	75.6	0.40007	59.4	0.40005	52.1	0.40014
20	73.7	0.40007	61.4	0.40005	52.2	0.40015
21	75.7	0.40007	59.3	0.40005	52.3	0.40015
22	73.6	0.40007	61.5	0.40005	52.4	0.40015
23	75.8	0.40007	59.2	0.40005	52.5	0.40015
⋮	⋮	⋮	⋮	⋮	⋮	⋮
585	54.7	0.40210	119	0.40207	119	0.40207

Sumber: Data olah, 2023

Lampiran 3. Titik Knot dan Nilai GCV untuk 2 Titik Knot

Nilai GCV untuk Variabel Usia (x_1) disetiap Orde

No	Orde Linier			Orde Kuadratik			Orde Kubik		
	k_{11}	k_{12}	Nilai GCV	k_{11}	k_{12}	Nilai GCV	k_{11}	k_{12}	Nilai GCV
1	16	24	0.39769	6	18	0.39785	3	4	0.39874
2	16	25	0.39770	6	17	0.39785	3	5	0.39874
3	17	26	0.39771	5	18	0.39785	4	5	0.39874
4	16	23	0.39771	7	18	0.39785	2	5	0.39874
5	17	21	0.39773	7	17	0.39785	2	4	0.39874
6	17	18	0.39773	5	17	0.39786	5	1	0.39875
7	16	26	0.39773	6	19	0.39786	4	1	0.39875
8	2	17	0.39773	5	19	0.39787	2	3	0.39875
9	4	8	0.39774	4	18	0.39787	2	6	0.39875
10	16	22	0.39774	8	17	0.39787	3	6	0.39875
11	5	9	0.39775	8	18	0.39787	6	1	0.39875
12	3	12	0.39776	7	19	0.39787	3	1	0.39875
13	18	24	0.39776	7	16	0.39787	4	6	0.39875
14	18	23	0.39776	4	19	0.39788	2	1	0.39875
15	16	27	0.39777	6	16	0.39788	7	1	0.39875
16	17	27	0.39777	4	17	0.39788	2	7	0.39875
17	16	22	0.39774	3	18	0.39788	5	6	0.39875
18	17	20	0.39777	8	16	0.39788	3	7	0.39876
19	19	22	0.39777	9	17	0.39789	4	7	0.39876
20	19	23	0.39777	3	19	0.39789	8	1	0.39876
21	18	22	0.39778	5	16	0.39789	2	8	0.39876
22	19	24	0.39779	8	19	0.39789	5	7	0.39877
23	18	25	0.39779	9	18	0.39789	3	8	0.39877
⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮
3364	15	25	0.39779	59	58	0.40119	59	58	0.40119

Sumber: Data olah, 2023

Nilai GCV untuk Variabel Berat Badan (x_2) disetiap Orde

No	Orde Linier			Orde Kuadratik			Orde Kubik		
	k_{21}	k_{22}	Nilai GCV	k_{21}	k_{22}	Nilai GCV	k_{21}	k_{22}	Nilai GCV
1	15	21	0.37888	14	18	0.37807	13	14	0.37917
2	15	22	0.37890	14	17	0.37824	13	15	0.37922
3	15	23	0.37902	15	16	0.37875	12	15	0.37940
4	15	20	0.37907	13	19	0.37881	12	16	0.37959
5	15	24	0.37924	13	18	0.37882	12	14	0.38009
6	15	25	0.37937	14	19	0.37883	11	16	0.38013
7	15	26	0.37947	15	17	0.37888	14	15	0.38023
8	15	19	0.37951	14	16	0.37922	13	16	0.38025
9	15	27	0.37962	13	20	0.37938	11	17	0.38045
10	15	28	0.37988	13	17	0.37965	11	15	0.38054
11	15	29	0.38005	15	18	0.37979	12	17	0.38055
12	4	15	0.38008	14	20	0.38013	10	17	0.38125
13	15	18	0.38017	13	21	0.38021	11	18	0.38136
14	5	15	0.38023	14	15	0.38090	10	16	0.38136
15	13	15	0.38032	12	19	0.38102	13	12	0.38144
16	14	15	0.38035	12	20	0.38109	12	13	0.38144
17	3	15	0.38046	13	16	0.38110	11	14	0.38167
18	8	15	0.38046	13	22	0.38114	10	18	0.38169
19	9	15	0.38049	16	17	0.38145	13	17	0.38201
20	15	17	0.38051	12	21	0.38145	12	18	0.38206
21	6	15	0.38052	12	18	0.38146	10	15	0.38210
22	7	15	0.38053	15	19	0.38156	14	16	0.38227
23	12	15	0.38053	14	21	0.38157	9	17	0.38244
⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮
785	3	2	0.41723	29	28	0.41665	29	28	0.41665

Sumber: Data olah, 2023

Nilai GCV untuk Variabel Panjang Badan (x_3) disetiap Orde

No	Orde Linier			Orde Kuadratik			Orde Kubik		
	k_{31}	k_{32}	Nilai GCV	k_{31}	k_{32}	Nilai GCV	k_{31}	k_{32}	Nilai GCV
1	52	68	0.39881	51	52	0.39920	61	62	0.40009
2	52	69	0.39882	51	53	0.39924	60	62	0.40009
3	52	67	0.39888	50	52	0.39925	60	61	0.40009
4	52	70	0.39890	50	53	0.39925	60	63	0.40009
5	51	68	0.39891	52	53	0.39928	61	63	0.40009
6	51	69	0.39891	49	53	0.39929	59	62	0.40009
7	53	68	0.39892	50	54	0.39931	59	63	0.40009
8	53	69	0.39894	51	54	0.39932	59	61	0.40009
9	52	66	0.39895	49	54	0.39932	62	63	0.40009
10	52	71	0.39896	49	52	0.39933	60	64	0.40009
11	51	70	0.39897	48	53	0.39933	59	64	0.40009
12	53	67	0.39899	50	51	0.39934	58	63	0.40009
13	51	67	0.39899	48	54	0.39935	58	62	0.40009
14	51	71	0.39901	47	53	0.39936	58	64	0.40009
15	52	72	0.39902	47	54	0.39936	61	64	0.40009
16	53	70	0.39903	53	46	0.39936	59	60	0.40009
17	53	66	0.39906	54	46	0.39937	58	61	0.40009
18	51	72	0.39907	52	54	0.39937	59	65	0.40009
19	52	73	0.39907	48	52	0.39940	57	63	0.40009
20	52	65	0.39908	50	55	0.39940	57	64	0.40009
21	51	66	0.39909	49	55	0.39941	58	65	0.40009
22	53	71	0.39909	48	55	0.39942	60	65	0.40009
23	54	68	0.39909	47	55	0.39943	62	64	0.40009
⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮
5184	47	55	0.40217	117	115	0.40212	117	115	0.40211

Sumber: Data olah, 2023

Lampiran 4. Estimasi Model Regresi Logistik Spline *Truncated*

Estimasi Model untuk variabel Usia (x_1)

```
> splinelin(y,x1,16,24)
Call:
polr(formula = y ~ m[, 2] + m[, 3] + m[, 4], data = DATA_FIX,
    method = "logistic")

Coefficients:
     m[, 2]      m[, 3]      m[, 4]
-0.08775838  0.11932361 -0.06440173

Intercepts:
 Gizi Baik|Gizi Kurang Gizi Kurang|Gizi Lebih
            -0.2231746              0.7375644

Residual Deviance: 23104.32
AIC: 23114.32
> tabel_uji_parsial
                           Value Std. Error   t value      p value
x1                    -0.08775838 0.007283625 -12.048723 1.969780e-33
x1_1                   0.11932361 0.014958829   7.976801 1.501743e-15
x1_2                  -0.06440173 0.010462800  -6.155306 7.493280e-10
Gizi Baik|Gizi Kurang -0.22317455 0.081009113  -2.754931 5.870443e-03
Gizi Kurang|Gizi Lebih 0.73756437 0.081590930   9.039784 1.569819e-19
```

Estimasi Model untuk variabel Berat Badan (x_2)

```
> splinelinx2(y, x2, 15)

Call:

polr(formula = y ~ m[, 2] + m[, 3], data = DATA_FIX, method =
"logistic")


Coefficients:

    m[, 2]     m[, 3]

-0.2157339  1.1237612


Intercepts:

 Gizi Baik|Gizi Kurang Gizi Kurang|Gizi Lebih

           -0.93837523              0.07696182


Residual Deviance: 22233.05

AIC: 22241.05

> tabel_uji_parsial

                            Value  Std. Error    t value       p value

x2                     -0.21573390 0.006780359 -31.817473 3.710964e-222

x2_1                    1.12376118 0.033277259  33.769643 5.504913e-250

Gizi Baik|Gizi Kurang  -0.93837523 0.072378737 -12.964791  1.937568e-38

Gizi Kurang|Gizi Lebih  0.07696182 0.073061517   1.053384  2.921651e-01
```

Estimasi Model untuk variabel Panjang Badan (x_3)

```
> splinelinx3(y, x3, 52, 68)

Call:
polr(formula = y ~ m[, 2] + m[, 3] + m[, 4], data = DATA_FIX,
    method = "logistic")


Coefficients:
     m[, 2]      m[, 3]      m[, 4]
 0.33198726 -0.41158840  0.05075549


Intercepts:
 Gizi Baik|Gizi Kurang Gizi Kurang|Gizi Lebih
              16.76604                17.72476


Residual Deviance: 23214.20
AIC: 23224.20
> tabel_uji_parsial
                             Value   Std. Error      t value      p value
x3                      0.33198726 0.0018721831   177.326281 0.000000e+00
x3_1                   -0.41158840 0.0088008224   -46.767038 0.000000e+00
x3_2                    0.05075549 0.0079850997     6.356275 2.067043e-10
Gizi Baik|Gizi Kurang  16.76604137 0.0002195782 76355.660786 0.000000e+00
Gizi Kurang|Gizi Lebih 17.72476448 0.0202277491   876.259855 0.000000e+00
```

Estimasi Model Simultan dengan 1 Titik Knot

```
> spline1knot(y, x1, 9, x2, 15, x3, 74)

Call:

polr(formula = y ~ m[, 2] + m[, 3] + m[, 4] + m[, 5] + m[, 6] +

    m[, 7], data = DATA_FIX, method = "logistic")


Coefficients:

     m[, 2]      m[, 3]      m[, 4]      m[, 5]      m[, 6]      m[, 7]

-0.20033815  0.17324779  0.06592540  0.89757587  0.02331175 -0.06351317


Intercepts:

 Gizi Baik|Gizi Kurang Gizi Kurang|Gizi Lebih

             0.9557389              1.9938619


Residual Deviance: 21947.03

AIC: 21963.03

> tabel_uji_parsial

                             Value Std. Error   t value       p value

x1                     -0.20033815 0.03368861 -5.946761  2.734998e-09

x1_1                    0.17324779 0.03369030  5.142364  2.713035e-07

x2                      0.06592540 0.01967920  3.350005  8.081011e-04

x2_1                    0.89757587 0.03872084 23.180692 7.130727e-119

x3                      0.02331175 0.01112930  2.094629  3.620396e-02

x3_1                   -0.06351317 0.01239168 -5.125468  2.967988e-07

Gizi Baik|Gizi Kurang   0.95573892 0.51904366  1.841346  6.557088e-02

Gizi Kurang|Gizi Lebih  1.99386187 0.51906102  3.841286  1.223914e-04
```

Estimasi Model Simultan dengan 2 Titik Knot

```
> splinekuad(y, x1, 6, 18, x2, 14, 18, x3, 51, 52)

Call:
polr(formula = y ~ m[, 2] + m[, 3] + m[, 4] + m[, 5] + m[, 6] +
    m[, 7] + m[, 8] + m[, 9] + m[, 10] + m[, 11] + m[, 12] +
    m[, 13], data = DATA_FIX, method = "logistic")


Coefficients:
     m[, 2]      m[, 3]      m[, 4]      m[, 5]      m[, 6]      m[, 7]      m[, 8]      m[, 9]
-0.77979125  0.04690966 -0.03653321 -0.01165019  0.80021492 -0.03471581  0.30697734 -0.40197270
    m[, 10]     m[, 11]     m[, 12]     m[, 13]
-9.82598015  0.10145037 -0.36298527  0.26083689


Intercepts:
 Gizi Baik|Gizi Kurang Gizi Kurang|Gizi Lebih
              -236.1314               -235.0725


Residual Deviance: 21635.88
AIC: 21663.88
> tabel_uji_parsial
                             Value   Std. Error       t value       p value
x1                        0.22805528 0.0068532098  3.327715e+01 8.267402e-243
x12                      -0.03259555 0.0029774475 -1.094748e+01  6.832124e-28
x1_1                      0.03846351 0.0043020140  8.940814e+00  3.863142e-19
x1_2                     -0.00602264 0.0014166982 -4.251181e+00  2.126464e-05
x2                        0.92302050 0.0008009533  1.152402e+03  0.000000e+00
x22                      -0.04143867 0.0008915131 -4.648127e+01  0.000000e+00
x2_1                      0.35379700 0.0090202180  3.922267e+01  0.000000e+00
x2_2                     -0.38408213 0.0002791459 -1.375919e+03  0.000000e+00
```

x3	3.73773949	0.0020046784	1.864508e+03	0.000000e+00
x32	-85.15489264	0.0030587298	-2.783995e+04	0.000000e+00
x3_1	-0.41019169	0.0099118426	-4.138400e+01	0.000000e+00
x3_2	0.39641767	0.0101019317	3.924177e+01	0.000000e+00
Gizi Baik|Gizi Kurang	-476.48784266	0.0004899293	-9.725645e+05	0.000000e+00
Gizi Kurang|Gizi Lebih	-475.41796165	0.0226141341	-2.102305e+04	0.000000e+00

Lampiran 5. Akurasi Model Regresi Logistik Ordinal Spline *Truncated*

```
> library(caret)
> Prediksi_Prob <- predict(ModelTerbaik, DATA_FIX, type = "class")
> confusionMatrix(as.factor(Prediksi_Prob), DATA_FIX$`Status Gizi Balita (BB/TB)`)
```

```
Confusion Matrix and Statistics

          Reference
Prediction     1     2     3
         1  1614   336     0
         2   488 13219   276
         3     0   177  1490

Overall Statistics

               Accuracy : 0.9274
                 95% CI : (0.9235, 0.9312)
    No Information Rate : 0.7802
    P-Value [Acc > NIR] : < 2.2e-16

                  Kappa : 0.797

 Mcnemar's Test P-Value : NA

Statistics by Class:
```

	Class: 1	Class: 2	Class: 3
Sensitivity	0.7678	0.9626	0.84371
Specificity	0.9783	0.8025	0.98882
Pos Pred Value	0.8277	0.9454	0.89382
Neg Pred Value	0.9688	0.8582	0.98268
Prevalence	0.1194	0.7802	0.10034
Detection Rate	0.0917	0.7511	0.08466
Detection Prevalence	0.1108	0.7945	0.09472
Balanced Accuracy	0.8731	0.8826	0.91627

Lampiran 6. Biodata Penulis

CURRICULUM VITAE

A. Data Pribadi

1. Nama : Samsul Arifin
2. Tempat, tanggal lahir : Sungguminasa, 4 Juli 1997
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4. Kewarganegaraan : Warga Negara Indonesia
5. Nomor HP/Wa : +62 853-9716-3116
6. e-mail : samsul.arrr@gmail.com
7. Bidang/Ketertarikan : Pemodelan Statistika, Machine Learning

B. Riwayat Pendidikan

1. Tamat SMA tahun 2016 di SMA Negeri 1 Pallangga
2. Sarjana (S1) tahun 2020 di Universitas Hasanuddin Departemen Statistika Program Studi Statistika
3. Magister (S2) tahun 2023 di Universitas Hasanuddin Departemen Statistika Program Studi Statistika

C. Pekerjaan dan Riwayat Pekerjaan

1. Tentor Matematika, FOCUS-A Makassar
2. Customer Service, PT. Bank Rakyat Indonesia, Tbk

D. Karya ilmiah yang telah dipublikasikan

Arifin, S., Islamiyati, A., & Herdiani, E. T. (2023). Ability of Ordinal Spline Logistic Regression Model in The Classification of Nutritional Status Data. Commun. Math. Biol. Neuroscience, 83.

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